Suppose we have the continuous-time transition equation

$$dX_t = TX_t dt + R dW_t$$

where  $X_t \in \mathbb{R}^n$ ,  $T \in \mathbb{R}^{n \times n}$ ,  $R \in \mathbb{R}^{n \times m}$ , and  $dW_t$  is a *m*-dimensional standard Brownian motion. In expectation, we have that this system behaves according to

$$dX_t = TX_t dt$$

which corresponds to the linear ODE system<sup>1</sup>

$$\frac{dX_t}{dt} = TX_t,$$

yielding the standard solution

$$X_t = \exp((t - \tau)T)X_{\tau},$$

where  $X_{\tau}$  is some initial condition (though  $\tau$  need not be zero).

Our problem is to find a matrix Z which convert flows into stocks. Our states,  $X_t$ , represents the instantaneous state variable, so, for example, if output is a state, then  $X_t$  tracks the *rate* of output. Therefore, to acquire the total output procued, a *stock* variable, we must integrate the path of  $X_t$ . In the case that T is invertible, we have the analytical solution<sup>2</sup>

$$\left| \int_{\tau}^{S} X_{t} dt \right| = \left| T^{-1} \exp((t - \tau)T) X_{\tau} - T^{-1} \right|.$$

Note that we use absolute values here to allow for  $\tau > S$  in the case that we want to determine output by integrating backwards in time (i.e. I observe my current state and guess the stock of output by extrapolating backward rather than observing my past state and guessing its evolution forward from there).

However, T is not always invertible, and in test trials with HANK models, T won't be invertible. Therefore, we approximate it using the definition of exponential matrices. Given any matrix T, we have that

$$\exp((t-\tau)T) = \sum_{n=0}^{\infty} \frac{((t-\tau)T)^n}{n!},$$

$$X_t - X_0 = \int_0^t TX_s \, ds + \int_0^t R \, dW_s.$$

The second integral is an Ito integral, which is a martingale. With standard Brownian motion, it has mean zero, so in expectation, our equation becomes

$$\mathbb{E}[X_t - X_0] = \int_0^t TX_s \, ds,$$

which is the *integral* equivalent of a linear ODE (i.e. every ODE can be represented with integrals).

 $^{2}$ To develop intuition for this, it may be useful to study the scalar case, where the invertibility of T does not matter, as it will just be some real.

<sup>&</sup>lt;sup>1</sup>Formally, any stochastic differential equation (SDE) is an *integral* equation. In our case, our SDE is actually short hand for

where  $((t - \tau)T)^0 = I$ . We can integrate term by term by the dominated convergence theorem<sup>3</sup>, yielding

$$\int_{\tau}^{S} \exp((t-\tau)T) dt = \sum_{n=0}^{\infty} \int_{\tau}^{S} \frac{((t-\tau)T)^{n}}{n!}$$

$$= (S-\tau) + \frac{(S-\tau)^{2}T}{2!} + \frac{(S-\tau)^{3}T^{2}}{3!} + \frac{(S-\tau)^{4}T^{3}}{4!} + \cdots$$

$$= (S-\tau) \sum_{n=0}^{\infty} \frac{((S-\tau)T)^{n}}{n!}.$$

We can therefore write

$$\int_{\tau}^{S} \exp((t-\tau)T) X_{\tau} ds = \left(\int_{\tau}^{S} \exp((t-\tau)T) ds\right) X_{\tau} = \left((S-\tau) \sum_{n=0}^{\infty} \frac{((S-\tau)T)^{n}}{n!}\right) X_{\tau}.$$

<sup>&</sup>lt;sup>3</sup>Since we know the exponential matrix converges, we can bound the tail above by some constant and integrate the resulting *finite* sum of powers of T, plus some constant. Since this function is integrable and dominates  $\exp((t-\tau)T)$ , we can apply the Lebesgue dominated convergence theorem and integrate term by term.